

A digital intervention for capturing the real-time health data needed for epilepsy seizure forecasting: formative codesign and the usability study protocol (the ATMOSPHERE study)

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Abstract

Background: Epilepsy is a chronic neurological disorder affecting individuals globally, marked by recurrent and apparently unpredictable seizures that pose significant challenges, including increased mortality, injuries, and diminished quality of life. Despite advancements in treatments, a significant proportion of people with epilepsy continue to experience uncontrolled seizures. The unpredictability of these events has been identified as a major concern for people with epilepsy, highlighting the need for innovative seizure forecasting technologies.

Objective: The ATMOSPHERE study aims to develop and evaluate a digital intervention, using wearable technology and data science, that provides real-time, individualised seizure forecasting for individuals living with epilepsy. This paper reports the protocol for one of the workstreams focusing on the design and testing of a prototype to capture real-time input data needed for predictive modelling. The aims were/are to (1) collaboratively design the prototype (work completed) and (2) conduct an 'in-the-wild' study to assess usability and to refine the prototype (planned research).

Methods: This study employs a person-based approach to design and usability test a prototype for real-time seizure precipitant data capture. Phase 1 (work completed) involved co-design with individuals living with epilepsy and healthcare professionals. Sessions explored users' requirements for the prototype, followed by iterative design of low fidelity, static prototypes. Phase 2 (planned research) will be an 'in-the-wild' usability study involving the deployment of a mid-fidelity, interactive prototype for four weeks, with the collection of mixed-methods usability data to assess the prototype's real-world application, feasibility, acceptability, and engagement. This phase involves primary participants (adults diagnosed with epilepsy) and, optionally, their nominated significant other. The usability study will run in three waves of deployment and data collection, aiming to recruit five participants per wave, with prototype refinement between waves.

Results: The ATMOSPHERE study aims to make a significant step forward in epilepsy management, focusing on the development of a user-centred, non-invasive wearable device for seizure forecasting. Through a collaborative design process and comprehensive usability testing, this research aims to address the critical need for predictive seizure forecasting technologies, offering a promising approach to improving the lives of individuals with epilepsy. By leveraging predictive analytics and personalised machine learning models, this technology seeks to offer a novel approach to managing epilepsy, potentially improving clinical outcomes, including quality of life through increased predictability and seizure management.

Conclusions: The ATMOSPHERE study represents a significant step forward in epilepsy management, focusing on the development of a user-centred, non-invasive wearable device for seizure forecasting. Through a collaborative design process and comprehensive usability testing, this research aims to address the critical need for predictive seizure forecasting technologies,

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offering a promising approach to improving the lives of individuals with epilepsy. By leveraging predictive analytics and personalised machine learning models, this technology seeks to offer a novel approach to managing epilepsy, potentially improving patients' clinical outcomes, including quality of life through increased predictability and control over their seizures.

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Original Manuscript

A digital intervention for capturing the real-time health data needed for epilepsy seizure forecasting: formative codesign and the usability study protocol (the ATMOSPHERE study)

ABSTRACT

<u>Background:</u> Epilepsy is a chronic neurological disorder affecting individuals globally, marked by recurrent and apparently unpredictable seizures that pose significant challenges, including increased mortality, injuries, and diminished quality of life. Despite advancements in treatments, a significant proportion of people with epilepsy continue to experience uncontrolled seizures. The apparent unpredictability of these events has been identified as a major concern for people with epilepsy, highlighting the need for innovative seizure forecasting technologies.

Objectives: The ATMOSPHERE study aims to develop and evaluate a digital intervention, using wearable technology and data science, that provides real-time, individualised seizure forecasting for individuals living with epilepsy. This paper reports the protocol for one of the workstreams focusing on the design and testing of a prototype to capture real-time input data needed for predictive modelling. The aims were/are to (1) collaboratively design the prototype (work completed) and (2) conduct an 'in-the-wild' study to assess usability and to refine the prototype (planned research).

Methods: This study employs a person-based approach to design and usability test a prototype for real-time seizure precipitant data capture. Phase 1 (work completed) involved co-design with individuals living with epilepsy and healthcare professionals. Sessions explored users' requirements for the prototype, followed by iterative design of low fidelity, static prototypes. Phase 2 (planned research) will be an 'in-the-wild' usability study involving the deployment of a mid-fidelity, interactive prototype for four weeks, with the collection of mixed-methods usability data to assess the prototype's real-world application, feasibility, acceptability, and engagement. This phase involves primary participants (adults diagnosed with epilepsy) and, optionally, their nominated significant other. The usability study will run in three waves of deployment and data collection, aiming to recruit five participants per wave, with prototype refinement between waves.

<u>Results:</u> The phase 1 co-design study engaged 22 individuals, resulting in the development of a mid-fidelity, interactive prototype based on identified requirements, including the tracking of evidence-based and personalised seizure precipitants. The upcoming Phase 2 usability study is expected to provide insights into the prototype's real-world usability, identify areas for improvement, and refine the technology for future development. The estimated completion date of Phase 2 is the last quarter of 2024.

<u>Conclusions:</u> The ATMOSPHERE study aims to make a significant step forward in epilepsy management, focusing on the development of a user-centred, non-invasive wearable device for seizure forecasting. Through a collaborative design process and comprehensive usability testing, this research aims to address the critical need for predictive seizure forecasting technologies, offering a promising approach to improving the lives of individuals with epilepsy. By leveraging predictive analytics and personalised machine learning models, this technology seeks to offer a novel approach to managing epilepsy, potentially improving clinical outcomes, including quality of life through increased predictability and seizure management.

KEYWORDS:

Epilepsy; Seizure forecasting; Data Science; Artificial intelligence; Machine learning; Wearable technology

INTRODUCTION

Background of the study

Epilepsy, a chronic neurological disorder, is characterised by recurrent and unpredictable seizures, affecting individuals across all age demographics. [1] As a leading cause of chronic morbidity globally, epilepsy

presents significant challenges, including a heightened risk of premature death, injury, and diminished quality of life (QoL). [2] With a prevalence rate of 5 to 10 cases per 1000 people, [1]the disorder imposes a considerable burden on individuals and healthcare systems alike. The financial impact is also notable, with the UK National Health Service (NHS) spending approximately £1.5 billion annually on care that ranges from primary treatment and management to addressing seizure-related injuries and mental health issues. [3], [4] The disability-adjusted life years (DALYs) rates in the United Kingdom, estimated at 92,400 per 100,000 population, [5] highlight the extensive health loss attributed to epilepsy. Sudden Unexpected Death in Epilepsy (SUDEP) remains a critical concern, with an incidence rate of approximately 1 in 1,000 people with epilepsy, annually, emphasizing the need for ongoing research and innovative management strategies. [6], [7] Accurate forecasting of seizures could allow people with epilepsy (PWE) to take additional safety precautions at times of increased risk.

Rationale for the study

Despite advancements in treatment – such as resective surgery, neurostimulation devices, new antiseizure medication (ASM), many with novel modes of action, not to mention the increasing availability of treatment across the developing world, with several ASMs on the World Health Organization's list of essential medicines – approximately one-third of the over 50 million people living with epilepsy globally still experience uncontrolled and apparently unpredictable seizures. [8] A large community survey (n=1056) [5] and the UK Epilepsy Priority Setting Partnership [9] ranked the unpredictability of seizures as one of the most impactful aspects of the condition and a research priority, underscoring an urgent need for solutions that can offer predictability and management of seizure risk. Seizure forecasting technologies emerge as a promising avenue in this regard, potentially enabling individuals to take pre-emptive measures during high-risk periods, therefore mitigating the risk of injury and premature death, and alleviating the psychological burden associated with the unpredictability of seizures. [10]

One potential avenue for seizure forecasting is leveraging technologies which directly monitor brain activity, combining this with data analytics to explore seizure risk modelling. Video-electroencephalography (v-EEG) is the gold standard for monitoring brain biosignals to detect seizure activity. However, this is not well suited to explore as an out-of-clinic seizure forecasting solution, as it is not feasible for long-term use in real-world scenarios. [11] Ultra-long-term EEG monitoring (UNEEG) represents a significant advancement in the field. [12] UNEEG's technology, involving surgical implantation, offers continuous EEG data capture in the user's real-world context. PWE who have trialled the technology have found the continual monitoring valuable, both in terms of internalising their locus of control and providing information to inform treatment changes. However, UNEEG is a minimally- invasive method with intrinsic risks and costs of surgery, making it inaccessible to a wider range of users. [12] [13] Non-invasive scalp EEG represents a more accessible solution that is designed for use outside of clinical settings. Yet, the adoption of this technology is hampered by limitations such as the stigma of visible devices, practicality, acceptance among users [13] and the technical and practical challenges of long-term outpatient monitoring. [14]

An alternative seizure forecasting approach is to explore predictive analytics based on seizure precipitants that can be captured via non-invasive, non-medical grade wearable devices, such as smartwatches. This presents a high-potential solution, as ubiquitous wearable technologies can be easily integrated into the enduser's everyday life, to continuously capture real-time data over the long term. Non-invasive wearable devices that can be used outside of clinical settings have been successfully used for epilepsy management in the form of the Empatica E4 wristband. [11] This device has shown promise in detecting epileptic seizures by

monitoring physiological parameters such as heart rate and acceleration. However, the Empatica E4 focuses primarily on seizure detection rather than forecasting. [11]

Exploration of seizure forecasting based on seizure precipitant data remains in its infancy. Emerging research has explored seizure cyclicity for forecasting models. However, these approaches on their own have limited accuracy for seizure forecasting. [15], [16], [17] There is potential to build upon and improve these models by including a greater range of seizure precipitants. This is a promising approach, given the existing evidence on the range of external and physiological variables that play a precipitating role in seizure propagation. Measurable biomarkers, such as emotional stress, infections, menstrual cycles, sleep deprivation, and alcohol use, have all been identified as seizure precipitants. The relationship between precipitants and seizure propensity could be quantified using mathematical models [17] [13], with these models informing algorithms to analyse person-specific data and facilitating personalised seizure forecasts.

Ubiquitous technologies with multimodal sensors, combined with ecological momentary assessment (EMA repeated sampling of an individual's cognitions, emotions, symptoms, and behaviours) could enable the real-time capture of the broad spectrum of known seizure precipitants. This real-time data, combined with predictive analytics, offers a feasible, accessible solution for seizure forecasting. To our knowledge, there is a pressing need to address the research gap in developing and evaluating technology for real-time seizure precipitant data collection, essential for supporting seizure forecasting interventions.

Person Based Approach

A technology to collect real-time seizure precipitant data relies on PWE actively engaging with the technology. As such, it is essential that a user-centred design approach is taken. The Person-based approach (PBA) [18]integrates methods from user-centred design and qualitative research to design and evaluate complex digital health technologies. This approach can lead to solutions which are more acceptable, feasible, meaningful, and optimally engaging to the people who will use them. The PBA advocates undertaking iterative design cycles with users through both formal qualitative research and patient and public involvement (PPI). Previous research [16], [19]underscores the necessity of employing PBA for the future design of epilepsy-related technologies, emphasizing iterative design cycles and user involvement to create solutions that are not only technologically effective but also deeply aligned with the needs and experiences of those living with epilepsy. Therefore, this study employed the PBA approach, recognizing its critical role in developing effective, user-cantered epilepsy management technologies.

The ATMOSPHERE study, funded by the Engineering and Physical Sciences Research Council (EPSRC) and in collaboration with the N-CODE network, [20] represents an effort to leverage technology for the benefit of individuals with epilepsy. The ATMOSPHERE project is an early-stage, proof-of-concept research initiative seeking to develop and evaluate a non-invasive, wearable epileptic seizure forecasting technological solution. This innovative approach aims to tackle the apparently unpredictable nature of seizures by collecting real-time data on various seizure precipitants, such as sleep quality, heart rate, menstrual cycle, medication adherence, and stress levels, from adults living with epilepsy.

The ATMOSPHERE project

The entire project is organised into several work packages, each targeting a specific aspect of seizure forecasting technology development:

Work Package 1: synthesising the evidence base on seizure precipitants or triggers.

Work Package 2: Designing and usability testing a mid-fidelity interactive prototype to capture real-time precipitant data from individuals living with epilepsy.

Work Package 3: Developing predictive models for seizure risk forecasting.

Work Package 4: Identifying user requirements for the end-user app interface including presenting seizure forecasting output. Developing a high-fidelity interactive prototype based on the outcome of the usability study in work package 2.

This protocol paper specifically addresses Workstream 2 of the ATMOSPHERE study, reporting the results from the completed co-design study and outlining the protocol for the upcoming usability study.

Work Package 2 aims

This multi-phase research workstream aims to design, develop, and test prototypes for collecting real-time data on seizure and seizure precipitants.

Objectives

- 1. Collaboratively design a mid-fidelity interactive prototype capable of collecting real-time seizure and seizure precipitant data from PWE.
- 2. Conduct in-the-wild usability testing of this prototype to evaluate its feasibility and user acceptance and engagement. This step is critical to assessing the data capture prototype's real-world application and identifying areas for improvement based on user feedback.
- 3. Use insights gained from usability testing to refine and enhance the prototype. Focus on improving its functionality, ease of use, and overall effectiveness in collecting high-quality data for seizure forecasting.

METHODS

Study design

Drawing on the Person Based Approach, [21]we undertook iterative PPI to codesign the data collection prototype (Phase 1). We are now preparing to conduct a mixed-methods 'in-the-wild' usability study in Phase 2 to further investigate the prototype's usability and implement refinements. (Figure 1)

PHASE 1: CO-DESIGN - COMPLETED WORK

Co-design contributors

Lived experience contributors: We disseminated an advert for the codesign sessions via Epilepsy Action (a UK epilepsy charity) and the SHAPE network (the Epilepsy Research Institute UK's patient and public involvement and engagement service). The advert invited individuals with lived experience of epilepsy to take part in the co-design workshops and provided a link to an expression of interest form (EoI) capturing consent for the research team to contact the individual. We invited individuals completing the EoI to complete a form capturing clinical and demographic characteristics – this data was pseudo-anonymised and used to understand the diversity of the group. We offered members the equivalent of £20 per hour in shopping vouchers.

Healthcare professional contributors: We used our existing professional networks to invite healthcare professionals with experience in providing care for epilepsy, both in primary and secondary care services.

Co-design sessions

Lived experience contributor sessions: The codesign involved iterative workshops, based on workshop plans and lasted 60-90 minutes. Early sessions explored the acceptability of collecting evidence-based precipitant data, views on the content of EMA items, the acceptability of different wearables for data collection, and the wider socio-technical space within which the technology would be used. From these early sessions, a guiding principles document [21]as developed to identify the users' needs and context and required components for the data-collection technology (presented in the findings below). Based on the guiding principles, we developed a low-fidelity prototype in Microsoft PowerPoint. In subsequent user consultation sessions, we iteratively refined the prototype by gathering feedback, completing a table of changes document (a summary of required changes) [21] and implementing required changes.

Health professional contributor sessions: We conducted one-to-one discussions online using a semistructured topic guide lasting 30-60 minutes. Discussions explored the health professionals' views on design considerations from the PWE's perspective, design considerations from the clinician's perspective (e.g. how data might fit into the clinical workflow), and an investigation into the potential harms or unintended consequences. From these discussions, a guiding principles document was developed which informed prototype design.

Prototype

At the end of the codesign process and in preparation for the usability study, an interactive mid-fidelity prototype was developed, see the results section for full details (Figure 2).

PHASE 2: USABILITY TESTING - PROTOCOL FOR PLANNED RESEARCH

Participants

Primary Participants: The primary participants in this study will be adults diagnosed with epilepsy. We will disseminate an advert via Epilepsy Action, the SHAPE network, and Caafi Health (an organisation with expertise in engaging underserved communities (e.g. those from ethnic minority groups) [22]. The inclusion and exclusion criteria for participation in the study are detailed in Table 1. These criteria were established to ensure that participants will be representative of the broader population of individuals with epilepsy, while also considering safety and ethical considerations pertinent to conducting research within this group.

Nominated Participants: The primary participants (adults living with epilepsy) will be invited to nominate a significant other, such as a carer, friend, or family member, to also participate in the study. This will be optional. The inclusion and exclusion criteria for nominated participants are detailed in Table 1.

Table 1: inclusion and exclusion criteria

Inclusion	Exclusion
Primary participant (participant with epilepsy	
diagnosis)	
Adult (over 18 years)	Those who have a major cardiac condition
Have a diagnosis of epilepsy	Those with a diagnosis of a major psychiatric
	condition (defined as needing secondary care

mental health services support)

Likely to have a seizure within the 4-week testing

Period (based on self-report)

Are fluent in English (to be able to engage with the

technology and qualitative interview)

Can provide consent

Can wear a wrist wearable device (smartwatch)

Able to input data into a smartphone app

Own a smartphone that can download apps from either the Apple store (iPhone) OR the Google Store (Android phones) (individuals were not required to own a smartwatch, as this was supplied as part of the study)

Nominated participants (a significant other identified

by the primary participant)

Nominated by a person meeting the criteria above

Adult (over 18 years)

Fluent in English (to be able to engage with the

technology and qualitative interview).

Able to consent to the study

Own a smartphone which can download apps from the

Google Play store or the Apple store

Be able to input data into a smartphone application

Recruitment and consent process

Primary Participants (individuals living with epilepsy): Potential participants will be given a participant information sheet and invited to contact the research team to express interest in the study. Those expressing interest will be asked to complete a sampling form, providing demographic and clinical information, this will be used to undertake purposeful sampling, aiming to recruit a diverse sample in terms of age, gender, ethnicity, and the type of epilepsy. Individuals sampled for the study will initially be contacted by an epilepsy clinician who will screen for eligibility. Eligible participants will then be contacted by a member of our research team for a recruitment consultation. Following the recruitment and screening consultation, informed consent will be obtained from individuals wishing to participate. This consent process will be conducted in accordance with ethical guidelines to ensure that participants are fully informed and willing to participate in the study voluntarily.

Nominated Participants: On entering the study, the primary will be invited to nominate a significant other (e.g. family member, friend, carer), and provide consent for us to contact this individual. We will contact these nominated individuals, provide them with a participant information sheet, undertake a recruitment consultation, and obtain fully informed consent from those wishing to take part.

Sampling and sample size

To achieve a comprehensive understanding of the usability and effectiveness of the intervention, we aim to recruit approximately 15 primary participants (with additional nominated participants). Recruitment will be organised into three waves, with each wave consisting of five participants. A sample size of five primary participants per wave was based on the usability testing literature, where 3-5 participants has been found to detect the majority of usability problems[23]. This phased approach will allow for iterative refinements to the intervention and study procedures based on feedback from each wave of participants, enhancing the robustness and relevance of our findings.

Study procedures

Participant procedure 1 - customisation survey and onboarding: the prototype is a smartwatch and smartphone app for the primary participant and a smartphone app only for the nominated participant (see methods section for a full description of the prototype). Participants will be sent a questionnaire (Using Microsoft Forms) to gather their requirements for customisation (see methods section for a description of customisation options), and the prototype will then be adapted to their requirements. Participants will be sent the prototype, along with details for downloading the smartphone app, and we will undertake an onboarding meeting to orientate participants to the technology.

Participant procedure 2 – deployment of data collection prototype: Participants will be invited to use the technology for four weeks. The four-week study period was intentionally chosen to encompass a complete menstrual cycle, enabling comprehensive capture of physiological data parameters across the full cycle. If the research team notice that the user has not worn the device or manually entered data, the participant will be sent an email offering a technical support meeting.

Participant procedure 3 - qualitative interview: At the end of the four-week usability period, the participants will be invited to a qualitative interview with a member of the research team (online or face-to-face, depending on participant preference and feasibility of travel for the research team). This qualitative interview will use a semi-structured topic guide, will be designed to last up to 60 minutes and will be audio recorded via Microsoft Teams. We will explore the following:

- Section 1: an open-ended exploration of the participant's views and experiences of the technology, including an exploration of the technology within the socio-technical space.
- Section 2: exploration of the technology features, e.g. co-tracking and passive data collection
- Section 3: exploration of the specific items and wording of the EMA
- Section 4: exploration of views to improve the technology (also gathered throughout sections 1-3)

Upon completion of the study, participants will receive financial compensation and can opt to receive a visual summary of the data collected.

Outcome measures to investigate feasibility, acceptability and engagement

The outcome measures being used to determine feasibility, acceptability and engagement are reported in Table 2.

Table 2: Outcome measures for feasibility, acceptability, and engagement

Outcome measure	Data points	Analysis
Quantitative outcome measures		
Smartwatch wear time	Assessed using the	Descriptive statistics ^a on
	presence of the heart	(1) Proportion of days device worn for
	rate data	(2) Number of hours during the day the
		device worn from
EMA response rate	Assessed for each EMA	Descriptive statistics on response rate for
	item	each EMA item
Qualitative outcome measure		
Users' perspectives	Qualitative interview	(1) Thematic analysis [24] on pseudo

data	anonymised transcripts to identify
	issues related to acceptability,
	feasibility and engagement
(2)	Tables of change to identify
	requirements to optimise the
	prototype

^a Descriptive statistics will include measures of central tendency (mean, median) and dispersion (standard deviation, interquartile range) to provide a comprehensive overview of the wear time/ response rate and missing data.

Prototype refinement

After each usability wave, we will analyse qualitative interview data to produce a tables of change document, identifying requirements to optimise the prototype [21]. If feasible, these changes will be implemented before the next wave of deployment (Figure 1).

Data management and ethical approval

Phase 1 involved a co-design with Patient and Public Involvement (PPI) contributors, which doesn't necessitate formal ethical approval. [25]However, we adhered to the General Data Protection Regulation (GDPR) guidelines. [26]To protect PPI contributor privacy, we used pseudo anonymisation for sampling forms, and we secured consent to report aggregated data. During co-design sessions, no direct quotes were captured, instead, we created summaries in the form of guiding principles and tables of changes.

For Phase 2, ethical approval was obtained from the University of Bristol, Faculty of Engineering Research Ethics Committee on March 10, 2022 (reference number: 10152). In alignment with our commitment to data privacy and protection, we carried out a Data Protection Impact Assessment (DPIA). [27] This assessment, ensures compliance with the GDPR within the European Union, [26]aims to evaluate and minimise potential risks to the rights and freedoms associated with personal data. The Information Compliance Manager & Data Protection Officer from the University of Bristol approved the DPIA on January 29, 2024.

RESULTS

PHASE 1: CO-DESIGN - COMPLETED WORK

We involved 12 individuals with lived experience of epilepsy, undertaking 21 consultations across the iterative design process. We involved 10 clinicians, with each individual contributing once. The demographic and clinical demographics are presented in Table 3. Table 4 presents the guiding principles derived from the codesign sessions. The first row is based on existing evidence, the subsequent rows are based on the codesign.

Table 3: Summary of numbers, demographic and clinical characteristics of PPI contributors

Characteristics	n (%)
Individuals with lived experience (n=12)	

Gender	Female	8 (67%)
	Male	4 (33%)
Geographical region	South East England	7 (58%)
	South West England	3 (25%)
	West Midlands	1 (8%)
	East Midlands	1 (8%)
Ethnicity	White British	10 (83%)
	Not reported	2 (17%)
Age	65 and over 55-64	1 (8%)
	45-54	2 (17%)
	35-44	2 (17%)
	25-34	4 (33%)
	23 34	1 (8%)
	Not reported	2 (17%)
Levels of deprivation ^a	Not reported 10% & 20% most deprived	2 (17%)
Levels of deprivation ^a		
Levels of deprivation ^a	10% & 20% <u>most deprived</u>	0
Levels of deprivation ^a	10% & 20% <u>most deprived</u> 30% & 40% most deprived	0 4 (33%)
Levels of deprivation ^a	10% & 20% most deprived 30% & 40% most deprived 50% most deprived	0 4 (33%) 1 (8%)
Levels of deprivation ^a	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived	0 4 (33%) 1 (8%) 3 (25%)
Levels of deprivation ^a Seizure types ^b	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived 30% & 40% least deprived	0 4 (33%) 1 (8%) 3 (25%) 4 (33%)
	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived 30% & 40% least deprived 10% and 20% least deprived	0 4 (33%) 1 (8%) 3 (25%) 4 (33%) 0
	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived 30% & 40% least deprived 10% and 20% least deprived Tonic clonic	0 4 (33%) 1 (8%) 3 (25%) 4 (33%) 0 5
	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived 30% & 40% least deprived 10% and 20% least deprived Tonic clonic Focal impaired awareness	0 4 (33%) 1 (8%) 3 (25%) 4 (33%) 0 5 4
	10% & 20% most deprived 30% & 40% most deprived 50% most deprived 50% least deprived 30% & 40% least deprived 10% and 20% least deprived Tonic clonic Focal impaired awareness Absence	0 4 (33%) 1 (8%) 3 (25%) 4 (33%) 0 5 4 3

	Atonic	1
	Myoclonic	1
	Drop seizures	1
	Not reported	3
Health professionals (n=10)		
Professional background	General Practitioner (GP)	3 (30%)
	Epilepsy Nurse Specialist	3 (30%)
	Learning Disability Nurse / Psychiatrist	2 (20%)
	Junior Doctor	2 (20%)

^a Estimated using contributors' postcodes against the 2019 English Indices of Deprivation data

Table 4: Guiding principles for data collection technology

User characteristics	Key feature(s)
Individuals with epilepsy	
Individuals with epilepsy are likely to have precipitants to seizures, as	Features to enable tracking of a
documented in the evidence	range evidence-based precipitants: emotional states, stress, fatigue,
	sleep disturbance, phase of menstrual cycle, seizure cycles, medication usage, fever or illness.
 In addition to evidence-based precipitants, individuals experience: a wide range of other seizure precipitants, e.g. cognitive overwhelm and certain foods. 	Features to offer customised tracking of personal seizure precipitants and prodromal
 Auras & prodromal symptoms, e.g. speech, emotional and behavioural changes 	symptoms, including subjective experiences.
Precipitants and prodromal symptoms can be personal (i.e. specific to that individual), subjective (i.e. hard to capture with objective/sensing data) and sometimes the individual can lack insight in these variables (see co-tracking below).	
SUDEP is a major concern for users, and users would like SUDEP risk	There are existing apps for SUDEP
assessments. Some users already complete regular SUDEP risk	risk assessment [28]. Consideration
assessments (including on mobile apps).	should be given to how the apps
	could work together.
With assurances around privacy and security, then individuals are	Features to maximise trust, identity,
positive about providing data that may improve forecasting performance.	privacy and security in the collection, processing and storage of data.
Most users are positive about active/manual tracking, viewing this as	Features to passively track
necessary for subjective data points (e.g. mood and fatigue). Many users	precipitants, prioritising passive

^b Some contributors identified as having multiple characteristics within the domain

already actively track seizures and precipitants, e.g. via paper diaries and spreadsheets. Some users described tracking as giving as a sense of control/empowering.

However, active tracking, can be a burden and challenging due to memory impairment and reduced insight. There may be strong individual differences (see "individual differences" below) in desire/ability to actively track.

Epilepsy management often involves carers/families/ friends, and individuals with epilepsy would like the technology to reflect this.

Some individuals with epilepsy are fully reliant on carers to track precipitants. Some manage independently but still have periods when they need support from their significant other/s to track precipitants (e.g. during periods of epilepsy-related amnesia/ reduced consciousness, due to cognitive impairment and due to deceased personal insight).

Individuals still want a sense of independence and control over the tracking process.

Different seizure types present different types of risks to individuals and require different self-management approaches.

There are <u>individual differences</u> in the user group. Some which need particular attention:

- The variability in the motivation for actively tracking data.
- Vulnerable and digitally excluded groups need careful consideration. This may include older adults (with a higher prevalence of certain types of epilepsy, e.g. post-stroke seizure), those with an intellectual disability (who have higher prevalence of epilepsy and treatment resistant seizures), those who are socially disadvantaged (e.g. lower level of education, drug users, prisoner population and homeless populations).

where possible.

Features to enable manual tracking of trigger data, as an option for those willing and able.

Features to enable optional **co- tracking**; whereby significant others
(family, friends, carers) can also track
data on the primary users health
states. This feature must preserve
independence, giving the user
control over who tracks, what they
track and control over stopping cotracking.

Feature for individuals to **track different seizure types**, to enable forecasting to differentiate between seizure types.

Simplicity and accessibility are essential. The technology should draw upon HCI principles to achieve inclusive design [29]

Clinicians

Secondary care:

Out-of-clinic data is critical for diagnosis and treatment planning. However, out-of-clinic data is often limited or can be inaccurate. Features to enable tracking of data useful for clinical management

Fuller and accurate data that would be useful for diagnosis and management includes:

- seizure data: type, frequency, and duration
- post-ictal symptoms
- medication adherence
- medication side effects
- the individual precipitants

Prototypes and tables of change

Based on the guiding principles, a low-fidelity prototype was initially developed, and a table of changes was used to capture user requirements for further refinement. Based on the final tables of change, a mid-fidelity, interactive prototype was developed, described below, and presented in Figure 2 and Table 5. The mid-fidelity, interactive prototype, and study materials (customisation questionnaire, onboarding process) were tested with 3 PPI contributors (all PWE) before the implementation of the main usability study.

Primary participant - Data collection will take place through the mEMA (by Illumivu) app installed on participant's smartphones, connected to a Garmin VivoActive 5 smartwatch worn by participants as outlined in Figure 2. Quantitative data captured passively through the Garmin wearable (see Table 5) will be streamed to the mEMA app and relayed to an encrypted database hosted by Illumivu. Simultaneously, Garmin sensor data will be relayed to Garmin Connect and processed via Garmin's algorithms to generate compound measures (based on a combination of variables), such as stress level and sleep score, which will be streamed to Illumivu's backend via Garmin's Health API (see Figure 2). The use of Garmin's API enables access to the raw data from the smartwatch. These data are transmitted via Bluetooth directly from the wearable device to the mEMA app on the phone. From there, raw data is sent up to the illumivu servers to be integrated with incoming EMA data.

The EMA surveys that participants will receive in the mEMA app have been developed for participants to record data on seizures (type, timing, duration and severity), prodromal symptoms, postictal symptoms, and seizure precipitants (triggers). There will be three modes for EMA completion:

- 1. Daily notifications to the participant at a set time to prompt EMA survey completion.
- 2. Contingent notifications, based on Garmin data (for example, elevated stress levels) to prompt survey completion.
- 3. User-initiated data entry, where the users can access the app and enter the data on an ad hoc basis. Completed surveys will be uploaded to the secure mEMA backend where they can be accessed by permissioned members of the core research team.

Table 5: Data collection protocol from primary participant

		Data point	Data collection schedule
Garmin	/passive	Sleep score	Daily summary
data			
		Heart rate	Continuous
		Respiration rate	Continuous
		Pulse ox blood oxygen saturation	Continuous
		Stress score	Continuous
		Accelerometer	Continuous
		Ambient light	Continuous
		Step count	Continuous
Smartphone	App/	Seizure event (type, timing, duration and	Notification ^a :
EMA		severity)	Twice daily
			<u>User-initiated</u> ^b
		Prodromal symptoms	Notification:
			Once daily

Postictal sympto	ms	Notification:
		Once, if seizure event reported
Epilepsy medica	tion usage (routine and	Notification:
emergency)		once daily notification
		<u>User-initiated</u>
Emotional states	and emotional intensity	Notification:
		Twice daily
		<u>User-initiated</u>
		Contingent c:
		If elevated stress detected on Garmin
		(capped at 3 notifications daily)
Sleep duration a	nd quality	Notification:
		Once daily
Acute illness or i	nfection	Notification:
		Once daily
Customisable pre	ecipitant question	Notification:
		Once daily
Alcohol intake		Notification:
		Once daily
Menstrual cycle		One-off series of questions

^a Notification: smartphone notification pushed to user as a prompt for them to complete the EMA item

Customisable components

The following are elements which can be customised by the participant:

- **EMA notification timings**: EMA notifications will be sent between 8am and 8pm. Users can indicate times within this window during which they do not wish to receive notifications.
- Optional survey items: Users can opt in/out of questions on menstrual cycle and EMA items on alcohol usage.
- **Personalised EMA item**: Users will be asked if they would like to track an individual/ personal seizure precipitant, and if feasible, we will create a personalised EMA item to capture this data.

Nominated participant for co-tracking (optional): during the customisation process, the primary participant will select which EMA items they consent for the nominated participant to receive (options being medication usage, emotional states and intensity, sleep duration and quality, alcohol intake, seizure events, prodromal symptoms, and postictal symptoms). The nominated participant will use the mEMA app installed on their smartphone and will receive notifications as prompts to complete the EMA survey.

PHASE 2: USABILITY TESTING - PLANNED RESEARCH

The findings from Phase 2, focusing on usability and the refinement of the prototype, are expected to be available in August 2024.

DISCUSSION

Principal findings

This paper reports (a) the codesign process of a novel technology designed to collect real-time data needed

^b User-initiated: users able to open the app and complete the EMA item any time

^c Contingent: a push notification prompting EMA completing sent, based on data from the Garmin device

for seizure forecasting and (b) the protocol for the usability testing of the prototype. Our codesign identified guiding principles for the prototype, including:

- 1. The implementation of tracking features for both evidence-based and custom seizure triggers, SUDEP risk data, and data to support clinical management. These features are designed to maximise user trust, identity protection, privacy, and security in the handling of data.
- 2. Where feasible and appropriate, passive tracking methods should be used, with the option for users to log data actively or manually on potential seizure precipitants.
- 3. The introduction of co-tracking features, giving users the freedom and capability to invite significant others (such as caregivers, friends, or family members) to participate in tracking precipitants of seizure onset.
- 4. The accommodation of diverse user needs and preferences, ensuring the design is inclusive and considers individual differences.

Strengths and limitations

This project emphasised the co-design of the prototype, ensuring that the technology aligns with the preferences and requirements of its target community. We engaged with clinicians from a range of relevant clinical specialities. We also engaged with a diverse group of individuals living with epilepsy, varying in aspects such as gender, age, seizure type, and socioeconomic background (determined by indices of deprivation). Individuals living with epilepsy are a diverse group, and we acknowledge that, it will be important to continue to gather more views on the technology beyond the 15 individuals involved in the phase 1 co-design. A notable limitation of phase 1 is the lack of ethnic diversity among contributors and no individuals from the very lowest indices of deprivation. To remedy this, in Phase 2, we are working with a specialised organisation to actively engage and involve ethnic minority groups and those from more deprived areas. This approach aims to make the development more inclusive and the technology more representative of the community it serves.

The second phase of our study is designed to evaluate the usability of our technology in a naturalistic setting (in-the-wild). The prototype undergoing testing is currently focused exclusively on the capabilities related to data collection. Consequently, it does not yet incorporate the advanced features, such as predictive analytics for determining seizure risk scores and the provision of seizure forecasting outputs to end-users. These more sophisticated functionalities are currently under development within separate workstreams. It is hypothesised that the availability of seizure forecasting output would serve as the primary motivation for individuals to continuously use the technology over an extended period. Therefore, the current limitation of the prototype, in its inability to replicate this aspect of real-world application and motivation, is acknowledged. To offset this limitation, we have endeavoured to introduce an alternative form of motivation for participants by offering summaries of their data (data visualisations of seizure and precipitant data) at the conclusion of the study period.

The data collection protocol, including the type and frequency of prompts, was codesigned with individuals with lived experience. Subsequently, the prototype was tested with three PPI contributors, who did not raise concerns about the frequency of prompts. However, there is a possibility that the frequency of prompts could be burdensome when the technology is used in the wild over an extended period. During qualitative interviews, we will explicitly explore participant's views on the data collection protocol, and we will refine the protocol, where necessary, based on this feedback.

The prototype developed for this phase is characterised as a mid-fidelity, interactive prototype, which permits the testing of its core functionalities, such as passive data collection and EMA. Despite its

capabilities, this prototype is not without its limitations, including a user experience (UX) interface that may not meet high standards and the necessity for some manual operations. However, the adoption of an agile methodology for this phase offers significant advantages. This approach allows for rapid learning and adjustments based on real-world, in-the-wild use, aligning with the principles of technological probing. While acknowledging its current constraints, this phase is pivotal in facilitating immediate, invaluable insights into the usability and potential impact of the technology in everyday contexts, guiding subsequent developments and enhancements.

This phase of the study did not include individuals with intellectual disabilities, an important population given their increased risk of epilepsy. [31] Recognising the necessity of designing interventions that cater specifically to their needs, it is evident that this group requires a thorough investigation into their specialised design requirements (beyond the scope of this study). Such an exploration should be a priority for future research to ensure the development of more inclusive and effective solutions. Further, the study focuses on adults, and a separate investigation for paediatric epilepsies is needed.

Overall, this work will inform the development of a high-fidelity interactive prototype. The assessments of that prototype will be conducted through user experience walkthroughs and qualitative interviews to identify usability and accessibility challenges. The results from these upcoming assessments will be reported in a separate paper for work package 4.

Conclusions

The co-design presented here identified user requirements for technology to capture the real-time data needed for seizure forecasting. Requirements include the ability to: collect data passively where possible (with options for manual data input); collect custom/personal seizure precipitant data (in addition to evidence-based precipitant data); enable co-tracking, where family/carers can input data; and consider inclusive design (due to variation in user characteristics and needs). The planned research in phase 2 will usability test a mid-fidelity, interactive prototype developed to meet these requirements. This co-design and usability testing contributes to a broader research aim; to develop seizure forecasting technology, leveraging real-time data on seizure precipitants, and employing predictive analytics. This technology could significantly improve the quality of life for individuals with epilepsy. By providing insights into the likelihood of seizure occurrence, such technology could offer a new level of independence and safety, reduce the economic and healthcare burdens associated with epilepsy, and facilitate a more personalised and efficient use of healthcare resources for epilepsy management. This technology development will be explored through proof-of-concept research of developing a non-invasive, wearable device specifically designed for seizure forecasting, integrating machine learning models for personalised predictive analytics, and utilising smartphone applications for real-time feedback to end-users.

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Conflicts of Interest

RC is a data scientist at Neuronostics. Neuronostics develop technologies for the diagnosis, treatment, and

monitoring of neurological conditions, including epilepsy.

Abbreviations

QoL: quality of life

NHS: National Health Service

DALYs: disability-adjusted life years

SUDEP: Sudden Unexpected Death in Epilepsy

PWE: people with epilepsy ASM: antiseizure medication EEG: electroencephalography

WDs: wearable devices

EMA: ecological momentary assessment

PBA: Person-based approach

PPI: patient and public involvement

EPSRC: Engineering and Physical Sciences Research Council

GDPR: General Data Protection Regulation DPIA: Data Protection Impact Assessment

UX: user experience

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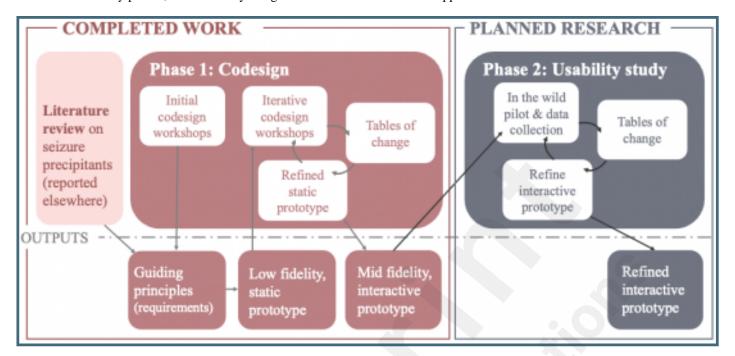
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Supplementary Files

Figures

Overview of study phases, with the study design based on the Person Based Approach.



An overview of the mid-fidelity, interactive prototype for the primary participant.

